

Uncoupling Accuracy and Stability in Tikhonov Regularization Parameter Selection

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Motivation

Ideal Problem: Solve $Ax = \bar{y} \in \text{Range}(A)$ for $x = x_{true}$

Actual Problem: Given $y = \bar{y} + \text{NOISE}$

Solve $Ax = y$

- ▶ X, Y - Hilbert spaces; $A : X \rightarrow Y$ **compact**, linear operator
- ▶ compact $\Rightarrow A^{-1}$ is "unbounded"



Shaw problem

Example - Fredholm Integral Equation of the 1st Kind
(ref. Shaw, Hansen)

$$\int_{-\pi/2}^{\pi/2} \kappa(t, \tau) x(\tau) dt = \bar{y}(\tau), \quad -\pi/2 \leq \tau \leq \pi/2,$$

$$\kappa(t, \tau) = (\cos(t) + \cos(\tau))^2 \left(\frac{\sin(\lambda)}{\lambda} \right)^2,$$

$$\lambda = \pi(\sin(t) + \sin(\tau))$$

Problem Specifications

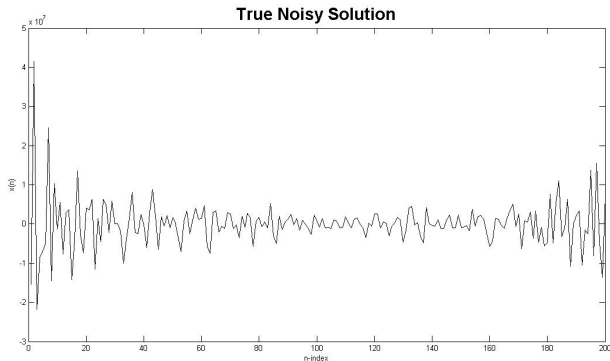
- ▶ $\bar{y} \in \mathbb{R}^{200}$, and x_{true} - sum of 2 Gaussian functions
- ▶ Condition number, $\kappa(A) \approx 5 \times 10^{19}$



Illustrative example

Noise $\approx 10^{-10}\%$

Approximate $x_{true} := A^{-1}\bar{y}$ by $x_{noise} := A^{-1}(\bar{y} + \text{NOISE})$



Noise completely contaminates solution!



Background:

Tikhonov-Lavrentiev regularization:

Fix $0 < \alpha < 1$ (for stability AND accuracy),

$$\text{Solve for } x_0: \quad (A + \alpha I)x_0 = y$$

Motivation for research efforts:

- ▶ (1948, 1963) A.N. Tikhonov, *Tikhonov Regularization*
- ▶ (1979) J.T. King, D. Chillingworth, *Iterated Tikhonov (model formulation, error bounds)*
- ▶ Current applications: Parameter estimation, turbulence modeling, imaging, etc.



Iterated Tikhonov Regularization

Algorithm

(Iterated Tikhonov-Lavrentiev regularization:)

- ▶ Fix $\alpha > 0$ (*for stability*)
- ▶ Fix $J > 0$ (*recover accuracy*)
- ▶ Solve for x_0 ,

$$(A + \alpha I)x_0 = y$$

- ▶ For $j = 1, 2, \dots, J$, solve for x_j

$$(A + \alpha I)(x_j - x_{j-1}) = y - Ax_{j-1}$$



Error estimation

Theorem

Fix $J > 0$. Fix $0 \leq \beta \leq J$. Let $\epsilon = \text{NOISE}$.

- ▶ **Regular solutions:** x_{true} in $\text{Range}(A^J)$ and $\|\epsilon\|_X \leq \epsilon_0$,

$$\|e_J\|_X \leq \frac{(J+1)\epsilon_0}{\alpha} + C(J)\alpha^{J+1}$$

- ▶ **Less regular solution:** x_{true} in $\text{Range}(A^\beta)$ and $\|A^{J-\beta}\epsilon\|_X \leq \epsilon_0$

$$\|e_J\|_X \leq \frac{(\beta+1)\epsilon_0}{\alpha} + C(\beta)\alpha^{\beta+1}$$

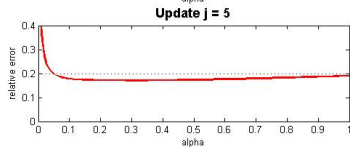
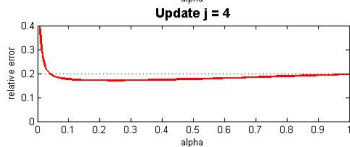
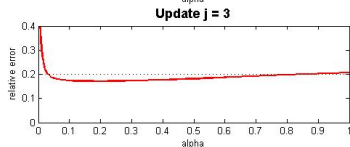
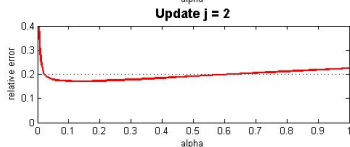
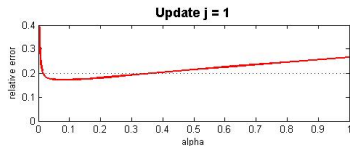
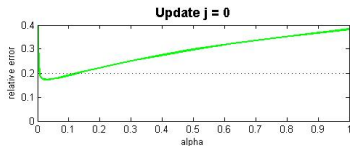
(Recover optimal convergence in large-scales)

$$\|A^{J-\beta}e_J\|_X \leq \frac{(J+1)\epsilon_0}{\alpha} + C(J)\alpha^{J+1}$$



Illustrative example

Noise, $\epsilon \approx 10^{-1}\%$



Descent properties

Let $x_{noise} := A^{-1}(\bar{y} + \text{NOISE})$, and $x_{true} = A^{-1}\bar{y}$

- ▶ x_{noise} minimizes $J_\epsilon(x) := \frac{1}{2}(Ax, x)_X - (\bar{y} + \text{NOISE}, x)_X$
- ▶ x_{true} minimizes $J_0(x) := \frac{1}{2}(Ax, x)_X - (\bar{y}, x)_X$

Theorem

- ▶ x_j : **minimizing sequence of $J_\epsilon(\cdot)$**
- ▶ Suppose $\|\epsilon\|_X \leq \epsilon_0$ and

$$\alpha \geq \frac{\epsilon_0}{\|x_{j+1} - x_j\|_X}, \text{ (stopping criterion)}$$

x_j : **"minimizing sequence" of noise-free functional $J_0(\cdot)$**



Parameter Sensitivity

Definition

The sensitivity of the iterated Tikhonov updates is

$$s_j(\alpha) = \frac{dx_j(\alpha)}{d\alpha}$$

Theorem

- ▶ Define $T_k(x_0(\alpha)) := x_0(\alpha) - \alpha x_0'(\alpha) + \dots + \frac{(-\alpha)^k}{k!} x_0^{(k)}(\alpha)$
- ▶ For NOISE = 0

$$x_{true} := x_0(\alpha = 0) = T_k(x_0(\alpha)) + O(\alpha^{k+1})$$

- ▶ For NOISE $\neq 0$

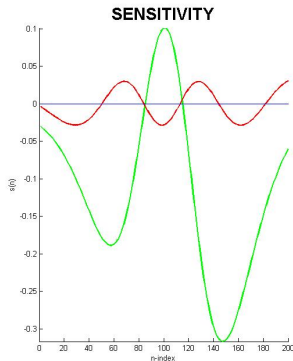
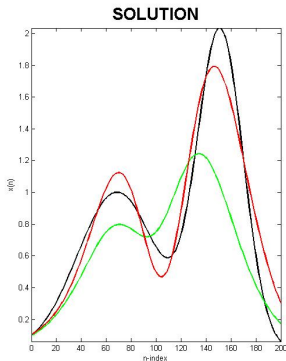
$$x_J(\alpha) = T_J(x_0(\alpha))$$



Numerical Results

$$\alpha = 1.00, J = 80$$

— = x_{true} , — = x_0 , — = x_J



$$s_0(\min) \approx -0.3167, \quad s_0(\max) \approx 0.1008$$

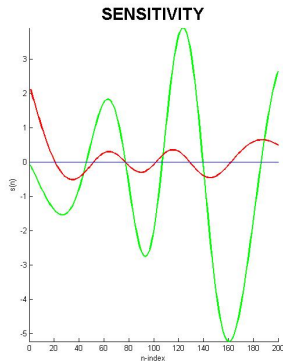
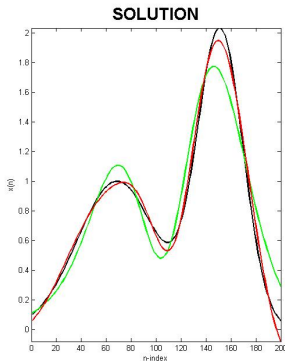
$$s_J(\min) \approx -0.0289, \quad s_J(\max) \approx 0.0301$$



Numerical Results

$$\alpha = 0.01, J = 80$$

— = x_{true} , — = x_0 , — = x_J



$$s_0(\min) \approx -5.2353, \quad s_0(\max) \approx 3.9044$$

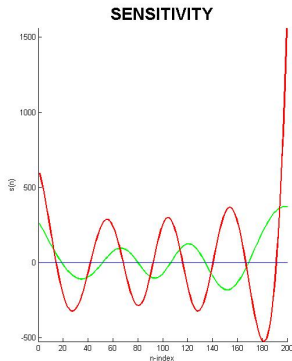
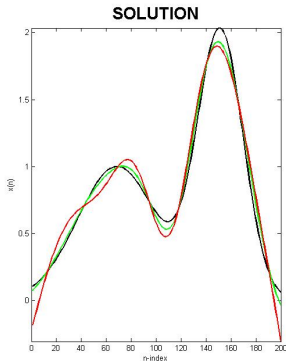
$$s_J(\min) \approx -0.5184, \quad s_J(\max) \approx 2.1215$$



Numerical Results

$$\alpha = 0.00001, J = 80$$

— = x_{true} , — = x_0 , — = x_J



$$s_0(\min) \approx -1.822E + 2, \quad s_0(\max) \approx 3.725E + 2$$
$$s_J(\min) \approx -5.252E + 2, \quad s_J(\max) \approx 1.562E + 3$$

